

Q 4

Q. 4 (i) $\nabla F(\bar{x}) = \begin{pmatrix} 20x_1 + 10x_2 + 4 \\ 10x_1 + 20x_2 - 10 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Q. 4 (ii) $= \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Eigen values of $\nabla^2 F(\bar{x})$

24.4536
 2
 -2.4536

$36 \text{ \& } 16$

at \bar{x}

Saddle point.

Local minima

Q 5

(i) FS subroutine
 \hat{x} tolerance NSC

(i) $N=5$ 0.92308 0.23077 8

(ii) $N=10$ 0.8958 0.02083 18

(iii) $N=20$ 0.88894 0.00017 38

(iv) $N=5$ 1.1538 0.23077 8

(v) $N=10$ 1.20833 0.020833 18

(vi) $N=20$ 1.2045 0.0001644 38

GS Subroutine

\hat{x} tolerance NSC

0.9929 0.200225 14

0.89278 0.020045 30

0.88898 0.00015 64

1.2126 0.2002258 14

1.2054 0.020045 30

1.2045 0.00015068



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	closed	bordered
(i)	✓	✓
(ii)	✓	✓

m
178885

2 / 7
(1.6, 0.8)



Q3 (i)

(b) ~~She~~ Sin is correct and Cortona's claim is false.

Q2

$$\beta = 1.623 \times 10^{-4} \quad 623$$

$$\epsilon = 1.623 \times 10^{-4}$$

S =

G

0.1 G

0.01 G

0.0001 G

9490804

949080323

9.49×10^{10}

9.49×10^{14}

Q. 2 (i)

30808

308072

30807148

Q. 2 (ii)

176

556

5551

Q. 2 (iii)

410

501

683

Q. 2 (iv)

9

9

10

Q. 2 (v)

9

$$k_e = \left\lceil \frac{\ln\left(\frac{1.623 \times 10^4}{0.5128}\right)}{\ln(0.975)} \right\rceil$$

$$= 319$$

$$k_{0.1e} = 410$$

$$k_{0.01e} = 501$$

$$k_{0.0001e} = \underline{\underline{683}}$$

(ii) $e_1 = 0.5$, $e_k = 1.9e_{k-1}^2$, $k \geq 2$,

using ~~Py~~ Python program,

$$k_e = 9$$

$$k_{0.1e} = 9$$

$$k_{0.01e} = 9$$

$$k_{0.0001e} = \underline{\underline{10}}$$

Is there a pen & paper
method available?

$$e_k = 0.5/k^2$$

$$k = \left\lceil \sqrt{\frac{0.5}{e_k}} \right\rceil$$

$$k_e = \left\lceil \sqrt{\frac{0.5}{1.623 \times 10^{-4}}} \right\rceil$$

$$= 56$$

$$k_{0.1e} = 176$$

$$k_{0.01e} = 556$$

$$\underline{\underline{k_{0.001e} = 5551}}$$

(iv) $e_1 = 0.5$

$$e_k = 0.975 e_{k-1}$$

$$\Rightarrow (E - 0.975) e_k = 0$$

$$\Rightarrow e_k = C_1 (0.975)^k$$

$$e_1 = C_1 \times 0.975 = 0.5$$

$$\Rightarrow C_1 = \frac{0.5}{0.975} = 0.5128$$

$$\Rightarrow \ln e_k = \ln C_1 + k \ln (0.975)$$

$$k = \left\lceil \frac{\ln\left(\frac{e_k}{C_1}\right)}{\ln(0.975)} \right\rceil$$

$$K_{0.01E} = \left\lceil \left(\frac{0.5}{0.01 \times 1.623 \times 10^4} \right) \right\rceil$$

$$\approx 9.49 \times 10^{10}$$

$$K_{0.0001E} \approx 9.49 \times 10^{14}$$

$$(ii) e_k = \frac{0.5}{K}$$

$$K = \left\lceil \frac{0.5}{e_k} \right\rceil$$

$$K_e = \left\lceil \frac{0.5}{1.623 \times 10^{-4}} \right\rceil = 3081$$

$$K_{0.1E} = 30808$$

$$K_{0.01E} = 308072$$

$$K_{0.0001E} = 30807148$$



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$$\beta = \underline{623}$$

$$\epsilon = 0.001\beta + 10^{-7}$$

$$= (0.001 \times 623 + 1) 10^{-7}$$

$$= \underline{1.623 \times 10^{-4}}$$

$$(i) e_k = 0.5 / \sqrt{k}$$

$$\sqrt{k} = 0.5 / e_k$$

$$k = \left[\left(\frac{0.5}{e_k} \right)^2 \right]$$

$$e_k < \epsilon$$

$$k_e = \left[\left(\frac{0.5}{\epsilon} \right)^2 \right]$$

$$= \left[\left(\frac{0.5}{1.623 \times 10^{-4}} \right)^2 \right]$$

$$= [9490803.222]$$

$$= 9490804$$

$$k_{0.1\epsilon} = \left[\left(\frac{0.5}{1.623 \times 10^{-5}} \right)^2 \right]$$

$$= 949080323$$

To find eigenvalues,

$$\begin{vmatrix} 32-\lambda & 8 \\ 8 & 20-\lambda \end{vmatrix} = (32-\lambda)(20-\lambda) - 64$$

$$= 640 - 52\lambda + \lambda^2 - 64$$

$$= \lambda^2 - 52\lambda + 576 = 0$$

$$\lambda = \frac{52 \pm \sqrt{52^2 - 4 \times 576}}{2}$$

$$= \underline{\underline{36 \text{ \& } 16}}$$

Point is on local minima

$$f(x) = 16x_1^2 + 8x_1x_2 + 10x_2^2 + 12x_1 - 6x_2 + 2$$

$$\nabla f(x) = \begin{bmatrix} 32x_1 + 8x_2 + 12 \\ 8x_1 + 20x_2 - 6 \end{bmatrix}$$

$$\nabla f(\bar{x} = [0.5, 0.5])$$

$$= \begin{bmatrix} -16 + 4 + 12 \\ -4 + 10 - 6 \end{bmatrix} = \underline{\underline{\begin{bmatrix} 0 \\ 0 \end{bmatrix}}}$$

$$\nabla^2 f = \begin{bmatrix} \frac{\partial}{\partial x_1} (32x_1 + 8x_2 + 12) \\ \frac{\partial}{\partial x_2} (32x_1 + 8x_2 + 12) \end{bmatrix}$$

$$\frac{\partial}{\partial x_1} (8x_1 + 20x_2 - 6)$$

$$\frac{\partial}{\partial x_2} (8x_1 + 20x_2 - 6)$$

$$= \begin{bmatrix} 32 & 8 \\ 8 & 20 \end{bmatrix}$$

To find eigen values.

$$\begin{vmatrix} 20-\lambda & 10 \\ 10 & 2-\lambda \end{vmatrix} = (20-\lambda)(2-\lambda) - 100 = 0$$

$$40 - 22\lambda + \lambda^2 - 100 = 0$$

$$\lambda^2 - 22\lambda - 60 = 0$$

$$\lambda = \frac{22 \pm \sqrt{22^2 + 240}}{2} = 24.4536, \\ \text{ \& } \\ -2.4536$$

Point on local minima is saddle point.



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Calculations

$$(1) f(\vec{x}) = 10x_1^2 + 10x_1x_2 + x_2^2 + 4x_1 - 10x_2 + 2$$

$$\nabla f = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \end{bmatrix} = \begin{bmatrix} 20x_1 + 10x_2 + 4 \\ 10x_1 + 2x_2 - 10 \end{bmatrix}$$

$$\nabla f(\vec{x} = (1.8, -4))$$

$$= \begin{bmatrix} 20 \times 1.8 - 40 + 4 \\ 18 - 8 - 10 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\nabla^2 f = \nabla(\nabla f^T)$$

$$\begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_1 \partial x_2} \\ \frac{\partial^2 f}{\partial x_2 \partial x_1} & \frac{\partial^2 f}{\partial x_2^2} \end{bmatrix} = \begin{bmatrix} \frac{\partial}{\partial x_1} (20x_1 + 10x_2 + 4) & \frac{\partial}{\partial x_1} (10x_1 + 2x_2 - 10) \\ \frac{\partial}{\partial x_2} (20x_1 + 10x_2 + 4) & \frac{\partial}{\partial x_2} (10x_1 + 2x_2 - 10) \end{bmatrix}$$

$$= \begin{bmatrix} 20 & 10 \\ 10 & 2 \end{bmatrix}$$

Checking Cortona's claim

~~co~~ f is coercive,

Coercive
 \Rightarrow

$$\lim_{\|x\| \rightarrow \infty} f(x) \rightarrow \infty$$

$$\|x\| \rightarrow \infty$$

$$\text{Let } p^T = [1 \ -1 \ 0]$$

$$q^T = [0 \ 0 \ -1]$$

$$\& \ x^T = [x_1 \ x_2 \ x_3] \mid x_1 \rightarrow x_2 \rightarrow x_3 \rightarrow \infty$$

Now ~~f is not~~
we can see f is not coercive

Have doubt about the definition & the way
I did

Seems Cortona is also right.

$$\begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 0 & 1 & 1 \end{bmatrix}$$

(i)

Given,

$$f(\bar{x}) = (p^T \bar{x})^2 + (q^T \bar{x})^2, \quad p, q, x \in \mathbb{R}^3$$

$$p \neq 0, q \neq 0, p+q \neq 0$$

Stk! $\exists \bar{x} \in \mathbb{R}^3$ with $\|\bar{x}\|_2 = 1$ & $f(\bar{x}) = 0$

Contra: f is coercive.

checking claim by Stk,

Take any \bar{x} , divide each component you will get \bar{x} with $\|\bar{x}\|$, you will get a vector with unit norm.

$$\text{take } p = [p_1 \ p_2 \ p_3]^T \quad \& \quad q = [q_1 \ q_2 \ q_3]^T$$

$$\& \quad \bar{x}^T = [x_1 \ x_2 \ x_3]$$

Note/ assume \exists

$$\begin{bmatrix} p^T \\ q^T \end{bmatrix} \begin{bmatrix} \bar{x} \end{bmatrix} = 0$$

$$= \begin{bmatrix} p_1 & p_2 & p_3 \\ q_1 & q_2 & q_3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = 0$$

We need 3 equations to find unique solution.

We can find multiple \bar{x} s satisfying $f(\bar{x}) = 0$ & divide each component by $\|\bar{x}\|$ to get a vector \bar{x} with unit norm \Rightarrow Claim by Stk is correct

$$\|A\|_F = \sqrt{a^2 + c^2 + 2b^2}$$

$$= \sqrt{a^2 + (1-a)^2 + 2b^2}$$

$$\|A\|_F^2 = 1 - 2a + 2b^2$$

$$2a^2 + 1 - 2a + 2b^2$$

To maximize $\|A\|_F^2$,

$$a(1-a) - b^2 \geq 0$$

$$\text{Make } a(1-a) = b^2$$

$$a = 0.5 \quad b = 0.5$$

$$\|A\|_F^2 = 1 - 2a + 2a(1-a)$$

$$= 1 - 2a + 2a - 2a^2$$

$$= 1 - 2a^2$$

$$\frac{d}{da} (2a^2 + 1 - 2a + 2a(1-a))$$

$$4a - 2 + 2 - 4a$$

$$2a^2 + 1 - 2a + 2a(1-a)$$

$$4a - 2 + 2 - 4a$$

Differentiating

$$-4a = 0 \Rightarrow a = 0$$

$$c = 1$$

$$\|A\|_F = 1$$

One sn Bounded

$$X = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$X = \begin{bmatrix} 0.5 & 0.5 \\ 0.5 & 0.5 \end{bmatrix}$$

(ii) Given,

$$S = \{ X \in \mathbb{R}^{2 \times 2} : X \text{ is symmetric \& +ve semi-definite, } \text{trace}(X) = 1 \}$$

Let A be an element in S .

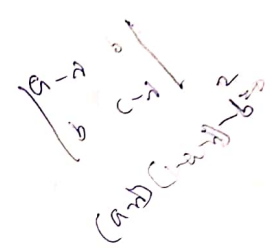
$$A = \begin{bmatrix} a & b \\ b & c \end{bmatrix} \quad \because \text{symmetric}$$

$$\begin{cases} a + c = 1 & \Rightarrow a + (1-a) = 1 \\ & 1 - a = c \\ ac - b^2 \geq 0 & \because \text{+ve semi-definite} \end{cases}$$

Existence of limiting point s
 $\Rightarrow \underline{\underline{S \text{ is closed}}}$

$$ac - b^2 \geq 0$$

* Assumption: Frobenius norm is $\leq M$ if the sets of matrices are bounded



This is the equation of a tilted ellipse

⇒ A is bounded

$$\underline{\underline{M = 1.78885}}$$

→ length of semimajor axis
Calculated using Wolfram Alpha

$$\underline{\underline{\hat{x} = (1.6, -0.8)}}$$

→ ~~the~~ one vertex of the ellipse.



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Q1

(D) Given,

$$Q \in \mathbb{R}^2$$

$$Q = \{ \bar{x} = (x_1, x_2) \in \mathbb{R}^2 : 9x_1^2 + 16x_1x_2 + 21x_2^2 \leq 16 \}$$

Q is closed since the limiting points ^{are} include the sets.

Testing if Q is bounded

$$f(\bar{x}) = 9x_1^2 + 16x_1x_2 + 21x_2^2$$

At minima/maxima

$$\nabla f(\bar{x}) = \begin{bmatrix} 18x_1 + 16x_2 \\ 16x_1 + 42x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$\Rightarrow \bar{x}^* = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

at (0,0), the function value is zero \Rightarrow ~~its min~~
(0,0) is minima since there exist \bar{x} where we
get bigger values.

For